

# Package ‘jmBIG’

March 21, 2024

**Type** Package

**Title** Joint Longitudinal and Survival Model for Big Data

**Version** 0.1.2

**Maintainer** Atanu Bhattacharjee <atanustat@gmail.com>

**Description** Provides analysis tools for big data where the sample size is very large. It offers a suite of functions for fitting and predicting joint models, which allow for the simultaneous analysis of longitudinal and time-to-event data. This statistical methodology is particularly useful in medical research where there is often interest in understanding the relationship between a longitudinal biomarker and a clinical outcome, such as survival or disease progression. This can be particularly useful in a clinical setting where it is important to be able to predict how a patient's health status may change over time. Overall, this package provides a comprehensive set of tools for joint modeling of BIG data obtained as survival and longitudinal outcomes with both Bayesian and non-Bayesian approaches. Its versatility and flexibility make it a valuable resource for researchers in many different fields, particularly in the medical and health sciences.

**Imports** Jmbayes2,joineRML,rstanarm,FastJM,dplyr,nlme,survival,ggplot2

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**Depends** R (>= 2.10)

**RoxygenNote** 7.3.1

**NeedsCompilation** no

**Repository** CRAN

**Author** Atanu Bhattacharjee [aut, cre, ctb],  
Bhriku Kumar Rajbongshi [aut, ctb],  
Gajendra K Vishwakarma [aut, ctb]

**Date/Publication** 2024-03-20 23:40:02 UTC

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cisurvfitJMCS	<i>Bootstrapped CI using FastJM</i>
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## Description

Bootstrapped CI for predicted survival probability

## Usage

```
cisurvfitJMCS(object)
```

## Arguments

object            a survfitJMCS object

## Value

Bootstrap CI for the survival probability and other relevant information for predicted survival plot

## Examples

```
##
library(survival)
library(dplyr)
jmcs1<-jmcsBig(dtlong=data.frame(long2),
dtsurv = data.frame(surv2),
longm=y~ x7+visit,
```

```

survm=Surv(time,status)~x1+visit,
rd= ~ visit|id,
samplesize=200,id='id')
mod2<-jmcs1
P2<-survfitJMCS(model<-mod2,ids<-c(5),u<-seq(surv2[surv2$id==5,]$time,
surv2[surv2$id==5,]$time+10,0.2),obs.time='time')
bootci<-cisurvfitJMCS(P2)
print(bootci)
##

```

---

jmbayesBig

*Joint model for BIG data using JMbayes2*


---

### Description

function for joint model in BIG DATA using JMbayes2

### Usage

```

jmbayesBig(
  dtlong,
  dtsurv,
  longm,
  survm,
  samplesize = 50,
  rd,
  timeVar,
  nchain = 1,
  id,
  niter = 2000,
  nburnin = 1000
)

```

### Arguments

dtlong	longitudinal dataset, which contains id,visit time,longitudinal measurements along with various covariates
dtsurv	survival dataset corresponding to the longitudinal dataset, with survival status and survival time
longm	fixed effect model for longitudinal response
survm	survival model
samplesize	sample size to divide the Big data
rd	random effect model part
timeVar	time variable in longitudinal model, included in the longitudinal data
nchain	number of chain for MCMC

id	name of id column in longitudinal dataset
niter	number of iteration for MCMC chain
nburnin	number of burnin sample for MCMC chain

**Value**

returns a list containing various output which are useful for prediction.

**Author(s)**

Atanu Bhattacharjee, Bhriгу Kumar Rajbongshi and Gajendra Kumar Vishwakarma

**References**

Rizopoulos, D., G. Papageorgiou, and P. Miranda Afonso. "JMbayes2: extended joint models for longitudinal and time-to-event data." R package version 0.2-4 (2022).

**See Also**

[jmcsBig](#), [jmstanBig](#), [joinRMLBig](#)

**Examples**

```
##
library(survival)
library(nlme)
library(dplyr)
fit5<-jmbayesBig(dtlong=long2,dtsurv = surv2,longm=y~ x7+visit,survm=Surv(time,status)~x1+visit,
rd= ~ visit|id,timeVar='visit',nchain=1,samplesize=200,id='id')
ydt<-long2>%filter(id%in%c(900))
cdt<-surv2[, 'id']>%filter(id%in%c(900))
newdata<-full_join(ydt,cdt,by='id')
P2<-predJMbayes(model<-fit5,ids<-c(900),newdata=newdata,process = 'event')
plot(P2$p1[[1]])
##
```

---

jmcsBig

*Joint model for BIG data using FastJM*

---

**Description**

function for joint model in BIG DATA using FastJM

**Usage**

```
jmcsBig(dtlong, dtsurv, longm, survm, samplesize = 50, rd, id)
```

**Arguments**

dtlong	longitudinal dataset, which contains id,visit time,longitudinal measurements along with various covariates
dtsurv	survival dataset corresponding to the longitudinal dataset, with survival status and survival time
longm	model for longitudinal response
survm	survival model
samplesize	sample size to divide the Big data
rd	random effect part
id	name of id column in longitudinal dataset

**Value**

returns a list containing various output which are useful for prediction.

**Author(s)**

Atanu Bhattacharjee, Bhriku Kumar Rajbongshi and Gajendra Kumar Vishwakarma

**References**

Li, Shanpeng, et al. "Efficient Algorithms and Implementation of a Semiparametric Joint Model for Longitudinal and Competing Risk Data: With Applications to Massive Biobank Data." Computational and Mathematical Methods in Medicine 2022 (2022).

**See Also**

[jmbayesBig](#),[jmstanBig](#),[joinRMLBig](#)

**Examples**

```
##
library(survival)
library(dplyr)
fit2<-jmcsBig(dtlong=data.frame(long2),dtsurv = data.frame(surv2),
longm=y~ x7+visit,survm=Surv(time,status)~x1+visit,rd= ~ visit|id,samplesize=200,id='id')
print(fit2)
##
```

---

`jmstanBig`*Joint model for BIG data using rstanarm*

---

**Description**

function for joint model in BIG DATA using rstanarm package

**Usage**

```
jmstanBig(  
  dtlong,  
  dtsurv,  
  longm,  
  survm,  
  samplesize = 50,  
  time_var,  
  id,  
  nchain = 1,  
  refresh = 2000  
)
```

**Arguments**

<code>dtlong</code>	longitudinal dataset, which contains id,visit time,longitudinal measurements along with various covariates
<code>dtsurv</code>	survival dataset corresponding to the longitudinal dataset, with survival status and survival time
<code>longm</code>	model for longitudinal response
<code>survm</code>	survival model
<code>samplesize</code>	sample size to divide the Big data
<code>time_var</code>	time variable in longitudinal model, included in the longitudinal data
<code>id</code>	name of id column in longitudinal dataset
<code>nchain</code>	number of chain for MCMC
<code>refresh</code>	refresh rate for MCMC chain

**Value**

returns a list containing various output which are useful for prediction.

**Author(s)**

Atanu Bhattacharjee, Bhriku Kumar Rajbongshi and Gajendra Kumar Vishwakarma

**References**

Goodrich, B., et al. "rstanarm: Bayesian applied regression modeling via Stan. R package version 2.17. 4." Online< <http://mc-stan.org> (2018).

**See Also**

[jmbayesBig](#), [jmcsBig](#), [joinRMLBig](#)

**Examples**

```
##
library(survival)
library(dplyr)
fit3<-jmstanBig(dtlong=long2,dtsurv = surv2,longm=y~ x7+visit+(1|id),
survm=Surv(time,status)~x1+visit,samplesize=200,time_var='visit',id='id')
P2<-postTraj(model<-fit3,m<-1,ids<-c(1,2,100))
pp1<-plot(P2$p1[[1]],plot_observed = TRUE)
pp2<-plot(P2$p1[[2]],plot_observed = TRUE)
pp3<-plot(P2$p1[[3]],plot_observed = TRUE)
##
```

---

joinRMLBig

*Joint model for BIG data using joineRML*


---

**Description**

function for joint model in BIG DATA using joineRML

**Usage**

```
joinRMLBig(dtlong, dtsurv, longm, survm, samplesize = 50, rd, timeVar, id)
```

**Arguments**

dtlong	longitudinal dataset, which contains id,visit time,longitudinal measurements along with various covariates
dtsurv	survival dataset corresponding to the longitudinal dataset, with survival status and survival time
longm	model for longitudinal response
survm	survival model
samplesize	random effect part
rd	random effect part
timeVar	time variable in longitudinal model, included in the longitudinal data
id	name of id column in longitudinal dataset

**Value**

returns a list containing various output which are useful for prediction.

**Author(s)**

Atanu Bhattacharjee, Bhriгу Kumar Rajbongshi and Gajendra Kumar Vishwakarma

**References**

Hickey, Graeme L., et al. "joineRML: a joint model and software package for time-to-event and multivariate longitudinal outcomes." *BMC medical research methodology* 18 (2018): 1-14.

**See Also**

[jmbayesBig](#), [jmstanBig](#), [jmcsBig](#)

**Examples**

```
##
library(survival)
library(dplyr)
fit4<-joinRMLBig(dtlong=long2,dtsurv = surv2,longm=y~ x7+visit,survm=Surv(time,status)~x1+visit,
rd=~ visit|id,timeVar='visit',samplesize=200,id='id')
P2<-predJRML(model<-fit4,ids<-c(10),dtlong=long2,dtsurv=surv2)
pp1<-plot(P2$plong[[1]])
pp1<-plot(P2$psurv[[1]])
##
```

---

long2

*longitudinal data*

---

**Description**

A longitudinal dataset with single marker , with different numeric and categorical covariate

**Usage**

```
data(long2)
```

**Format**

a tibble of 13 columns and 5639 observations,

**id** id value for subjects

**status** survival status

**time** survival time

**y** longitudinal marker

**visit** visit time of longitudinal measurements

**x1,x2,...,x7** different numeric and categorical variable

---

longsurv

*longitudinal- survival dataset*

---

**Description**

A longitudinal dataset with single marker , with different numeric and categorical covariate

**Usage**

```
data(longsurv)
```

**Format**

a tibble of 13 columns and 5639 observations,

**id** id value for subjects

**status** survival status

**time** survival time

**y** longitudinal marker

**visit** visit time of longitudinal measurements

**x1,x2,...,x7** different numeric and categorical variable

---

plot\_cisurvfitJMCS      *Plot for cisurvfitJMCS object*

---

### Description

prediction of survival probability and longitudinal marker using FastJM for BIG data

### Usage

```
plot_cisurvfitJMCS(object)
```

### Arguments

object                  fitted survfitJMCS object

### Value

Plot for predicted survival probability

### Examples

```
##
library(survival)
library(dplyr)
jmcs1<-jmcsBig(dtlong=data.frame(long2),
dtsurv = data.frame(surv2),
longm=y~ x7+visit,
survm=Surv(time,status)~x1+visit,
rd= ~ visit|id,
samplesize=200,id='id')
mod2<-jmcs1
P2<-survfitJMCS(model<-mod2,ids<-c(5),u<-seq(surv2[surv2$id==5,]$time,
surv2[surv2$id==5,]$time+10,0.2),obs.time='time')
P3<-cisurvfitJMCS(P2)
plot_cisurvfitJMCS(P3)
##
```

---

postSurvfit                  *Prediction using rstanarm*

---

### Description

posterior survival probability estimates from rstanarm for BIG data

**Usage**

```
postSurvfit(model, ids, ...)
```

**Arguments**

model	fitted model
ids	value of id
...	other parameter option, see posterior_survfit

**Value**

list of predicted value for the given id

**Examples**

```
##
library(survival)
library(dplyr)
jmstan<-jmstanBig(dtlong=long2,
  dtsurv = surv2,
  longm=y~ x7+visit+(1|id),
  survm=Surv(time,status)~x1+visit,
  samplesize=200,
  time_var='visit',id='id')
mod1<-jmstan
P2<-postSurvfit(model<-mod1,ids<-c(1,2,210))
pp1<-plot(P2$p1[[1]])
pp1
pp2<-plot(P2$p1[[2]])
pp2
pp3<-plot(P2$p1[[3]])
pp3
##
```

---

 postTraj

*Prediction using rstanarm*


---

**Description**

prediction of the posterior trajectory for longitudinal marker while using rstanarm for Big data

**Usage**

```
postTraj(model, m, ids, ...)
```

**Arguments**

model	fitted model object
m	m for posterior_traj function
ids	value of id
...	other parameter option, see posterior_traj

**Value**

list of predicted values for the given id

**Examples**

```
##
library(survival)
library(dplyr)
fit6<-jmstanBig(dtlong=long2,dtsurv = surv2,longm=y~ x7+visit+(1|id),
               survm=Surv(time,status)~x1+visit,samplesize=200,time_var='visit',id='id')
P2<-postTraj(model<-fit6,m<-1,ids<-c(1,2,100))
pp1<-plot(P2$p1[[1]],plot_observed = TRUE)
pp2<-plot(P2$p1[[2]],plot_observed = TRUE)
pp3<-plot(P2$p1[[3]],plot_observed = TRUE)
##
```

---

predJMbayer

*Prediction using JMbayer2*

---

**Description**

prediction of survival probability and longitudinal marker using jmBayes2 for BIG data

**Usage**

```
predJMbayer(model, ids, process = "longitudinal", newdata, ...)
```

**Arguments**

model	fitted model object
ids	value of id
process	see jm
newdata	dataset having covariate information for the ids mentioned above.
...	other parameter options, see predict.jm

**Value**

list of predicted value for the given id

**Examples**

```
##
library(survival)
library(nlme)
library(dplyr)
jmcs1<-jmbayesBig(dtlong=long2,
dtsurv = surv2 ,
longm=y~ x7+visit,
survm=Surv(time,status)~x1+visit,
rd= ~ visit|id,
timeVar='visit',
nchain=1,
samplesize=200,
id='id')
mod3<-jmcs1
ydt<-long2>%filter(id%in%c(900))
names(ydt)
cdt<-surv2[, 'id']>%filter(id%in%c(900))
names(cdt)
newdata<-full_join(ydt,cdt,by='id')
P2<-predJMbayes(model<-mod3,ids<-c(900),newdata=newdata,process = 'event')
plot(P2$p1[[1]])

##
```

---

 predJRML

*Prediction using joinerML*


---

**Description**

prediction of survival probability and longitudinal marker using joinerML for BIG data

**Usage**

```
predJRML(model, ids, dtlong, dtsurv)
```

**Arguments**

model	fitted model object
ids	value of id
dtlong	longitudinal data
dtsurv	survival data

**Value**

list of predicted values for the given id

**Examples**

```
##
library(survival)
library(dplyr)
jmcs1<-joinRMLBig(dtlong=long2,
dtsurv = surv2,
longm=y~ x7+visit,
survm=Surv(time,status)~x1+visit,
rd=~ visit|id,
timeVar='visit',
samplesize=200,
id='id')
mod4<-jmcs1
P2<-predJRML(model<-mod4,ids<-c(10),dtlong=long2,dtsurv=surv2)
plot(P2$plong[[1]])
plot(P2$psurv[[1]])
##
```

---

print

*print*

---

**Description**

print

**Usage**

```
print(object, ...)
```

**Arguments**

object	object
...	others

**Value**

prints table containing various parameter estimates, SE, P- value for both survival and longitudinal submodel, if the model is bayesian it includes their credible interval too.

**Examples**

```
##
library(survival)
library(dplyr)
fit7<-jmstanBig(dtlong=long2,dtsurv = surv2,longm=y~ x7+visit+(1|id),
survm=Surv(time,status)~x1+visit,samplesize=200,time_var='visit',id='id')
#####
fit8<-jmcsBig(dtlong=data.frame(long2),dtsurv = data.frame(surv2),
longm=y~ x7+visit,survm=Surv(time,status)~x1+visit,rd= ~ visit|id,
samplesize=200,id='id')
#####
fit9<-jmbayesBig(dtlong=long2,dtsurv = surv2 ,
longm=y~ x7+visit,survm=Surv(time,status)~x1+visit,
rd= ~ visit|id,timeVar='visit',nchain=1,samplesize=200,id='id')
#####

fit10<-joinRMLBig(dtlong=long2,dtsurv = surv2,
longm=y~ x7+visit,survm=Surv(time,status)~x1+visit,
rd=~ visit|id,timeVar='visit',samplesize=200,id='id')
```

---

```
print.jmbayesBig      print.jmbayesBig
```

---

**Description**

print method for class 'jmbayesBig'

**Usage**

```
## S3 method for class 'jmbayesBig'
print(object, digits = 4, ...)
```

**Arguments**

object	object
digits	used for round the numeric values after decimal
...	others

**Value**

prints table containing various parameter estimates, SE, P- value for both survival and longitudinal submodel, if the model is bayesian it includes their credible interval too.

**Examples**

```
##
library(survival)
library(dplyr)

#####
mod3<-jmbayesBig(dtlong=long2,
dtsurv = surv2 ,
longm=y~ x7+visit,
survm=Surv(time,status)~x1+visit,
rd= ~ visit|id,
timeVar='visit',
nchain=1,
samplesize=200,
id='id')
print(mod3)
```

---

```
print.jmcsBig
```

```
print.jmcsBig
```

---

**Description**

print method for class 'jmcsBig'

**Usage**

```
## S3 method for class 'jmcsBig'
print(object, digits = 3, ...)
```

**Arguments**

object	object
digits	used for round the numeric values after decimal
...	others

**Value**

prints table containing various parameter estimates, SE, P- value for both survival and longitudinal submodel, if the model is bayesian it includes their credible interval too.

**Examples**

```
##
library(survival)
library(dplyr)
```

```
#####
mod2<-jmcsBig(dtlong=data.frame(long2),
dtsurv = data.frame(surv2),
longm=y~ x7+visit,
survm=Surv(time,status)~x1+visit,
rd= ~ visit|id,
samplesize=200,id='id')
print(mod2)
```

---

```
print.jmstanBig      print.jmstanBig
```

---

## Description

print method for class 'jmstanBig'

## Usage

```
## S3 method for class 'jmstanBig'
print(object, digits = 3, ...)
```

## Arguments

object	object
digits	used for round the numeric values after decimal
...	others

## Value

prints table containing various parameter estimates, SE, P- value for both survival and longitudinal submodel, if the model is bayesian it includes their credible interval too.

## Examples

```
##
library(survival)
library(dplyr)
mod1<-jmstanBig(dtlong=long2,
dtsurv = surv2,
longm=y~ x7+visit+(1|id),
survm=Surv(time,status)~x1+visit,
samplesize=200,
time_var='visit',id='id')
print(mod1)
```

---

```
print.joinRMLBig      print.joinRMLBig
```

---

**Description**

print method for class 'joinRMLBig'

**Usage**

```
## S3 method for class 'joinRMLBig'
print(object, digits = 4, ...)
```

**Arguments**

object	object
digits	used for round the numeric values after decimal
...	others

**Value**

prints table containing various parameter estimates, SE, P- value for both survival and longitudinal submodel, if the model is bayesian it includes their credible interval too.

**Examples**

```
##
library(survival)
library(dplyr)
mod4<-joinRMLBig(dtlong=long2,
dtsurv = surv2,
longm=y~ x7+visit,
survm=Surv(time,status)~x1+visit,
rd=~ visit|id,
timeVar='visit',
samplesize=200,
id='id')
print(mod4)
```

---

surv2	<i>survival data</i>
-------	----------------------

---

**Description**

A survival dataset related the long2 dataset, with different numeric and categorical covariate

**Usage**

```
data(surv2)
```

**Format**

a tibble of 13 columns and 1000 observations,

**id** id value for subjects

**status** survival status

**time** survival time

**visit** visit time of longitudinal measurements

**x1,x2,...,x7** different numeric and categorical variable

---

survfitJMCS	<i>Prediction using FastJM</i>
-------------	--------------------------------

---

**Description**

prediction of survival probability using FastJM for BIG data

**Usage**

```
survfitJMCS(model, ids, u, method = "GH", obs.time)
```

**Arguments**

model            fitted model object

ids              value of id

u                see survfitjmcs

method          options are 'Laplace','GH'

obs.time        vector which represents time variable in the longitudinal data

**Value**

list of predicted value for the given id along with other information relevant for survival probability  
confidence plot

**Examples**

```
##
library(survival)
library(dplyr)
jmcs1<-jmcsBig(dtlong=data.frame(long2),
dtsurv = data.frame(surv2),
longm=y~ x7+visit,
survm=Surv(time,status)~x1+visit,
rd= ~ visit|id,
samplesize=200,id='id')
mod2<-jmcs1
P2<-survfitJMCS(model<-mod2,ids<-c(5),u<-seq(surv2[surv2$id==5,]$time,
surv2[surv2$id==5,]$time+10,0.2),obs.time='time')
print(P2)
##
```

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